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NONLINEAR MAPPING DESCRIBING LEARNING PROCESS HAVING MEMORY

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A nonlinear mapping model for learning and memory is proposed. The learning is a process that visual patterns presented in an information source(IS) cause selective activations of units composing a random network(RN). The process is a nonlinear map from the visual patterns onto the sets of the selective activations of the units. The aggregation is a set of the clusters in which the units are interconnected strongly with a coupling line, while the other units, not belonging to the clusters, are interconnected loosely. During a lapse of time, the states are assumed to be modified by two effects for the firing units; a rehearsal effect and a forgetting effect. The affected states show serial position curves(SPC) on which the points are arranged by a sequential list of the visual patterns presented.

We present the visual patterns, A_k ($k = 1, 2, \dots, n$), specified with occurrence probabilities $P(A_k)$ in the randomly arranged IS. Let X be an original element defined by

$$X = \begin{pmatrix} A_1, & A_2, & \dots, & A_n \\ P(A_1), & P(A_2), & \dots, & P(A_n) \end{pmatrix}, \quad (1)$$

where $\sum_{k=1}^n P(A_k) = 1$. The mapping is decomposed by successive ones $\varphi^{(k)}$ ($k = 1, 2, \dots, n$). The aggregation of the clusters yield a potential denoted by $y^{(k)}$. To define $y^{(k)}$, let $M(A_k)$'s be clusters and c_k be the number of the clusters $M(A_k)$. Now we introduce a normalized number of $M(A_k)$, that has a probability ρ_k , i.e. $\rho_k = c_k / \sum_{k=1}^n c_k$. Then the mapping is formally defined as follow

$$\varphi^{(k)} : X^{(k)} = \begin{pmatrix} A_k \\ P(A_k) \end{pmatrix} \longrightarrow Y^{(k)} = \begin{pmatrix} M(A_k) \\ \rho_k \end{pmatrix}. \quad (2)$$

In the mapping $\varphi^{(k)}$'s, we have not to take account of a time delay of $y^{(k)}$ explicitly. To evaluate the time delay, we divide the potential $y^{(k)}$ into its components $y_j^{(k)}$ and we regard $y^{(k)}$ as a vector potential. Expressed differently, we associate the elements $Y^{(k)}$'s ($k = 1, 2, \dots, n$) representing the aggregations with the vector potential $y^{(k)}$'s ;

$$\begin{aligned} y^{(k)} &= (y_1^{(k)}, y_2^{(k)}, \dots, y_k^{(k)}, 0, \dots, 0), \\ \longleftrightarrow Y^{(k)} &\equiv E^{(k-1)} \cdot Y^{(k-1)} + \varphi^{(k)}(X^{(k)}), \end{aligned} \quad (3)$$

where $E^{(k-1)}$ is a $(k-1) \times (k-1)$ matrix specifying a correlation between the mappings $\varphi^{(k-1)}$'s and the j -th components of $y^{(k)}$ are given by

$$y_j^{(k)} = \begin{cases} \sum_{j=1}^{k-1} e_{ij}^{(k-1)} y_j^{(k-1)} & (1 \leq i \leq k-1, 1 \leq j \leq k-1) \\ y_k^{(k)} & (j = k) \end{cases}, \quad (4)$$

where $e_{ij}^{(k)}$'s are the matrix elements. With the aid of the matrix element $e_{ij}^{(k)}$, we characterize the correlation by the effective coupling lines. During a lapse of time, the effective coupling lines between the clusters will be changed.

To evaluate the memory effects arising from the correlations, we suppose that $e_{ij}^{(k)}$'s are characterized by the two effects,

$$e_{ij}^{(k)} = R_{ij}^{(k)} + D_{ij}^{(k)}, \quad (5)$$

where $R_{ij}^{(k)}$ and $D_{ij}^{(k)}$ represent a rehearsal effect and a forgetting(decaying) effect, respectively.

We assume exponential dependences for $R_{ij}^{(k)}$ and $D_{ij}^{(k)}$;

$$R_{ij}^{(k)} = R_{ij}^{(0)} \exp(-r_0(ij) \{k - (i+j)/2\}^2), \quad (6)$$

$$D_{ij}^{(k)} = D_{ij}^{(0)} \exp(-d_0(ij) \{k - (i+j)/2\}^2), \quad (7)$$

for $i, j = 1, 2, \dots, k-1$ and $k = 1, 2, \dots, n$, in the numerical calculation.

For our purpose it is sufficient to know that the notations $R_{ij}^{(0)}$, $D_{ij}^{(0)}$, $r_0(ij) (< 0)$, $d_0(ij) (> 0)$ denote some values of constant.

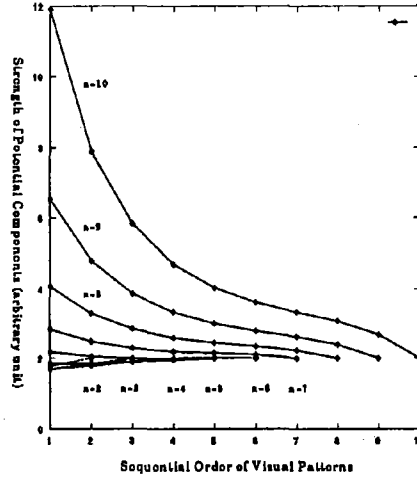


Figure 1: The behaviors of the errors show serial position curves(SPC). In the numerical calculations, we put $R_{ii}^{(0)} = 1.0$, $R_{ij}^{(0)} = 0.04$, $D_{ii}^{(0)} = -0.1$, $D_{ij}^{(0)} = -0.01$, and we also use the constants given as $r_0(ii) = 0.1$, $r_0(ij) = -0.005$, $d_0(ii) = 0.1$, $d_0(ij) = -0.005$, respectively.

Fig. 1 shows behaviors of variation for potential components, by which the errors would be interpreted.

As a result, we gave a nonlinear mapping model for memory and learning. The learning was a process expressed by the nonlinear map from IS onto RN. States of the RN were determined by the selective activations. Furthermore, we introduced aggregations of the clusters. During a lapse of time, the coupling line between the units or the effective coupling line between the clusters have changed, and eventually the changes have caused errors in a retrieval task. As curves showing the errors we obtained serial position curves(SPC).